Technische Universität München
Lehrstuhl Informatik V
M. Bader, A. Pöppl

WS 2018/19
Worksheet 3
November 28th 2018

Tutorial: HPC - Algorithms and Applications WS 18/19

Complete the following assignments (alone or in a group), and upload your solutions on *Moodle* until Sunday, December 9th 2018.

Worksheet 3: Coalesced Access, Sparse Linear Algebra

T3.1: Recap on Coalesced Access

Consider this CUDA kernel call:

```
dim3 grid(8, 8, 1); dim3 block(32, 32, 1);
kernel<<<grid, block>>>(A);


__global__ void kernel(float* A) {
  int tx = threadIdx.x, ty = threadIdx.y, tz = threadIdx.z;
  const int n = 64;
  float f;
  /* set value of f here */
}
```

For each of the following instructions, answer shortly if access to the array `A` is uncoalesced, partially coalesced or coalesced (chipset: NVidia Fermi, CUDA cc $\geq$ 2.0).

a) `f = A[tx];`

b) `f = A[(tx * n + ty) * n + tz];`

c) `f = A[tx + 1];`

d) `f = A[ty];`

e) `f = A[2 * tx];`

f) `f = A[n * tx];`

g) `f = A[tx / 2 + 16];`

h) `f = A[ty * tx];`

T3.2: CSR kernel

a) Write a CSR matrix-vector multiplication kernel for the PageRank example code.

- Define grid and block size in `kernels.cu`
- Implement the `k_csr_mat_vec_mm` kernel:
  - i) Assign a matrix row to each thread
  - ii) Compute the row $\times$ vector product in a loop.
  - iii) Add the result to the output vector.
- Compile the code using `make`. You will have to choose a suitable C compiler (gcc is the default in `Makefile`).

b) Try the kernel on a small matrix (`mtx/my.mtx`). The program output should be:

```
x_1 = 3.602992e-01
x_2 = 2.700142e-02
x_3 = 6.238353e-02
x_4 = 2.431707e-02
x_5 = 4.042290e-01
x_6 = 2.700142e-02
x_7 = 2.431707e-02
x_8 = 5.378454e-02
x_9 = 1.666666e-02
```

c) Next, test the kernel on a bigger matrix. Download `flickr.mtx`[1] or `usroads.mtx`[2] in the Matrix Market format (`*.mtx`) and run the PageRank algorith on the matrix (if it's too big choose a different matrix). Which page is the most relevant according to the algorithm?

[1] `https://www.cise.ufl.edu/research/sparse/matrices/Gleich/flickr.html`
[2] `https://www.cise.ufl.edu/research/sparse/matrices/Gleich/usroads.html`

H3.1: Vectorized CSR kernel

a) Write a vectorized CSR matrix-vector multiplication kernel for the PageRank example code.

- Implement the `k_csr2_mat_vec_mm` kernel in `kernels.cu`:

  i) Set grid and block size in the kernel call accordingly

  ii) Assign a matrix row to each *warp* now

  iii) Allocate a shared array `vals[]` for the partial results of a block

  iv) Compute one row × vector product in a loop. This time, parallelize the loop over all 32 threads in the warp. Take care that access to the arrays `indices` and `data` is coalesced.

  v) Use a reduction of some kind (ideally: binary fan-in) to add up the partial sums in `vals[]` and add the output to the result vector.

b) Try the new kernel on *mtx/my.mtx* and check if the output is consistent with T3.2b

c) Test both CSR kernels on the big matrix from T3.2c and measure execution times (`time ./sparse mtx/flicker.mtx`). How does performance compare?